



Integrating isolated trees improves the agricultural
performance assessment of smallholder farming systems at
landscape scale in the Senegalese peanut basin

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« Make our planet treed again »

(Montpellier Declaration, World Agroforestry Congress, 2019)



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Agroforestry parklands in face of SDG's

SUSTAINABILITY

EQUITY

RESILIENCE

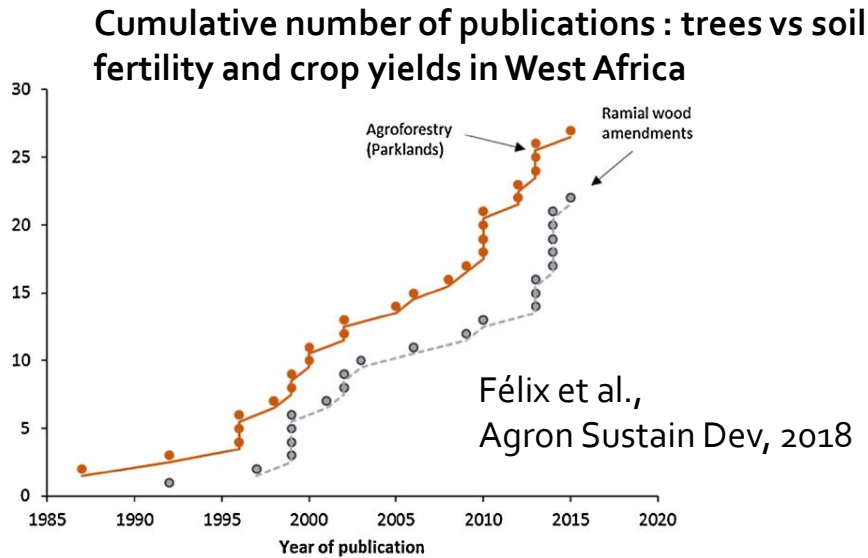


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Background & objectives



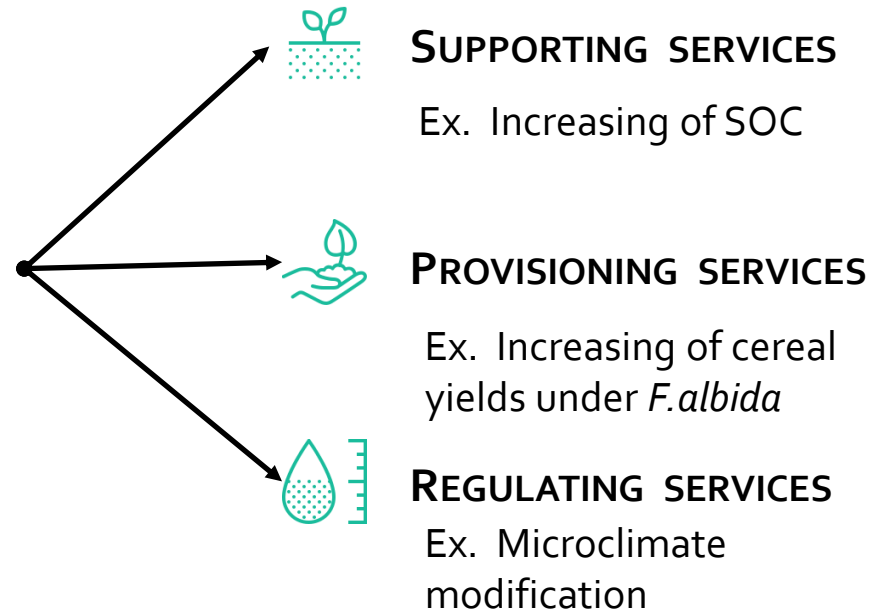
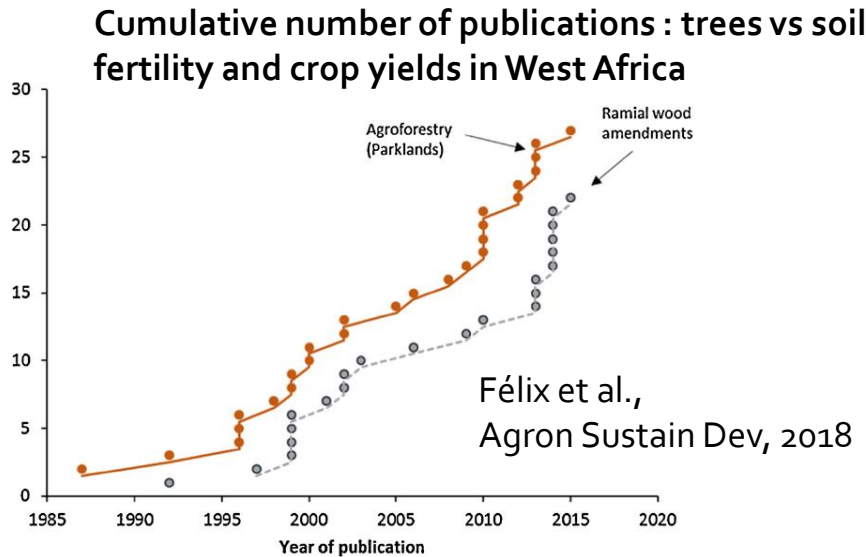
An increasing scientific interest in understanding effects of parklands on soil and crop productivity



Background & objectives



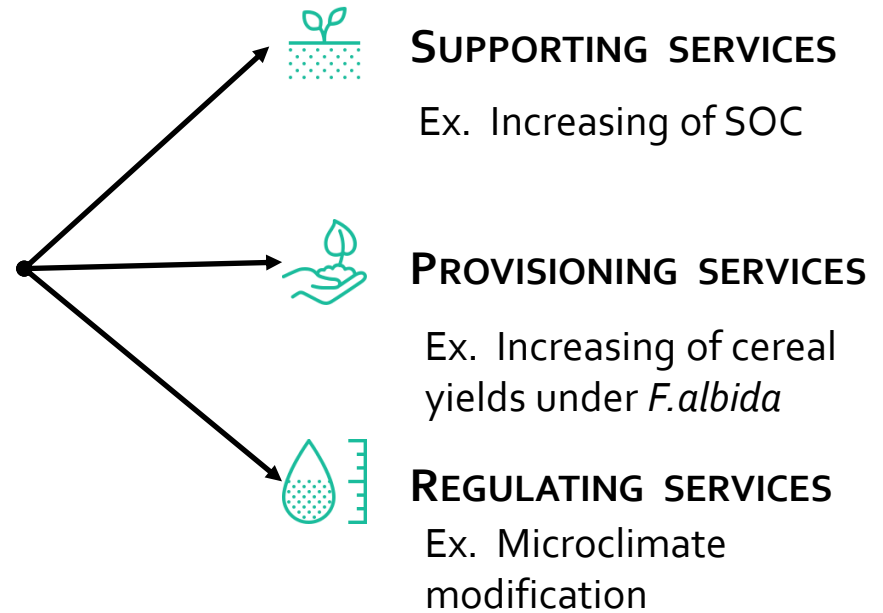
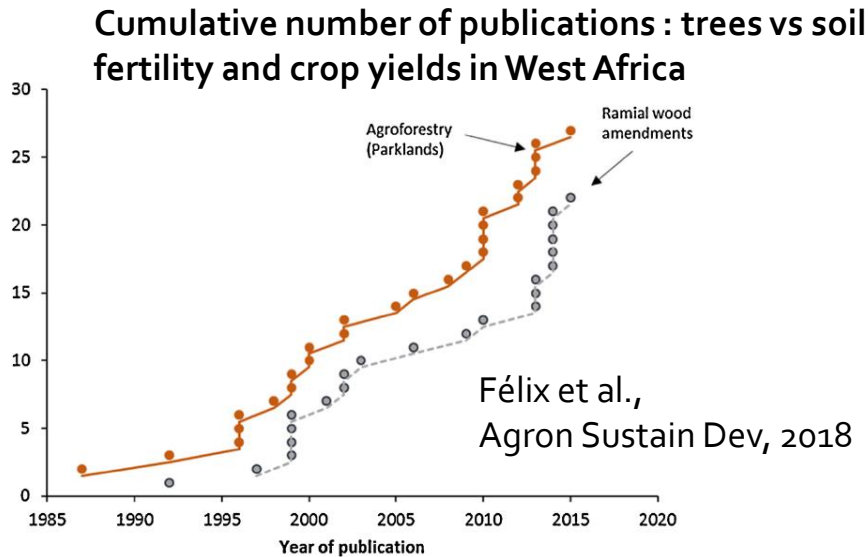
An increasing scientific interest in understanding effects of parklands on soil and crop productivity



Background & objectives



An increasing scientific interest in understanding effects of parklands on soil and crop productivity



... But still some challenges

- 1 – Most of studies are conducted **at tree scale**
- 2 – Limited knowledge on the impacts of **parklands pattern** (composition / structuring) on agricultural performance of farming system **at landscape scale**
- 3 – **Models** (crop process-based or statistical) accounting for trees in agricultural landscape remain scarce

Background & objectives





At landscape scale :

1 - To evidence the contribution of parklands structuring on the agricultural performance of smallholder farming systems

2 - To estimate crop yields and its spatial heterogeneity

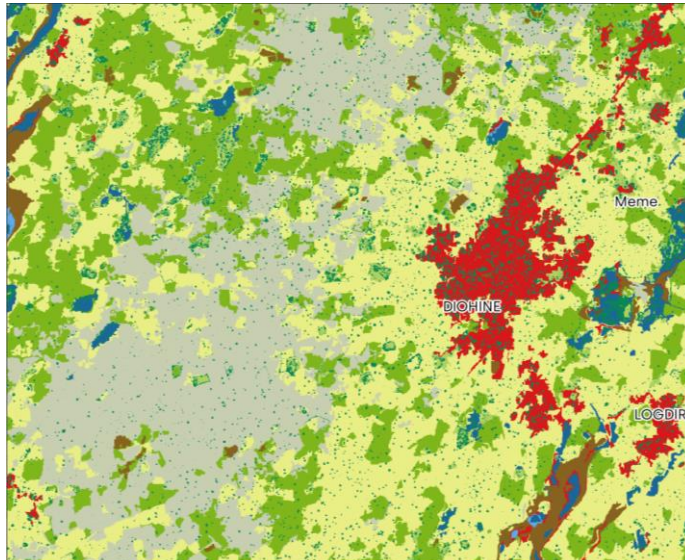
Where, What, How? Materials & Methods



The « old » peanut basin : the Senegalese breadbasket



A agricultural landscape dominated by rainfed crops ...

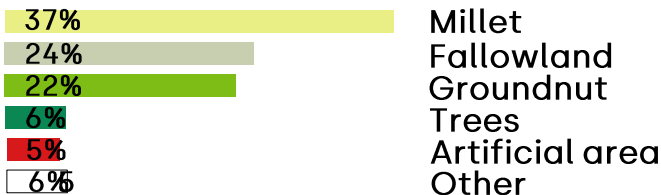


CLIMATE

- *Sudanian climate
- *Annual rainfall : **500-650 mm**
- *Rainy season : July to Oct.

FARMING SYSTEM

- * Agriculture dominated by:
 - **Millet** (on-farm consumption)
 - **Groundnut** (cash crop)
 - Livestock
- * **Low input**



Ndao et al., 2019

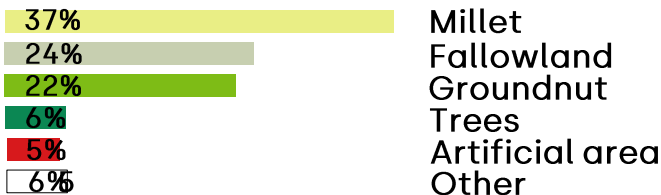
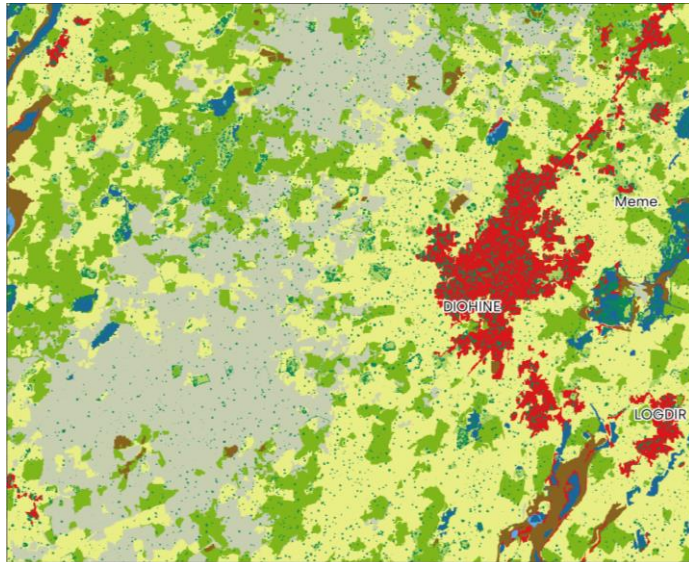


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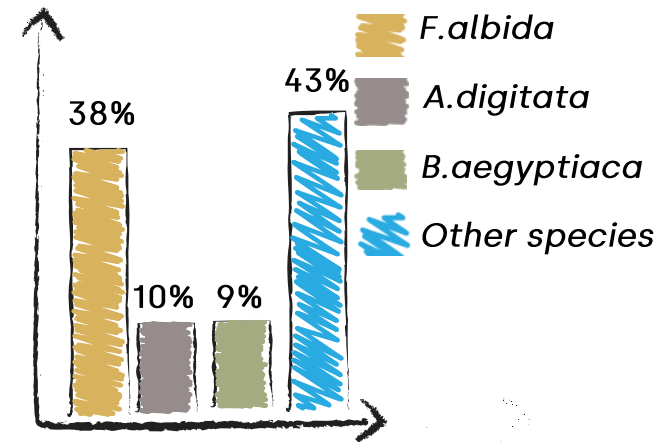
FARMING SYSTEM

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Ndao et al., 2019



... And a *F.albida* parkland



- *Leguminous nitrogen-fixing specie
- * 'reverse phenology'
- * 'fertility hotspot' at tree level
- *And various other tree species



Where, What, How? Materials & Methods



At the nexus of remote sensing, landscape ecology and statistical modelling

DATA

METHODS AND OUTPUTS



AGRONOMICAL SURVEY

[MILLET – 50 FIELDS]

[4 CLASSES OF LANDSCAPE]

[2 CROPPING SEASONS]



1.Agricultural
practices

2.Tree
inventory

3.Yield
components

Where, What, How? Materials & Methods



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Linear Mixed model

→ 1-TREE EFFECT

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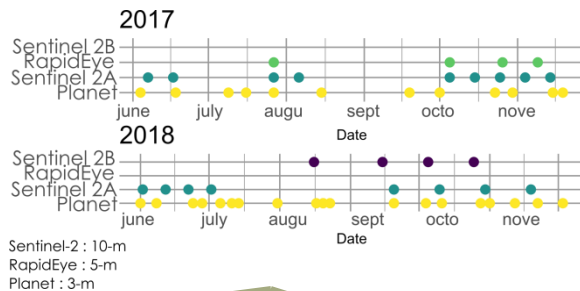
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→ 1-TREE EFFECT



MULTI-SOURCES REMOTE SENSING



1. Parkland structuring proxies

- *Nbs of trees
- *Woody cover
- *Tree density

2. Vegetation productivity proxies

- *Phenological metrics
- *Vegetation indices
- *Water stress index
- *Nutrient stress index

Where, What, How? Materials & Methods



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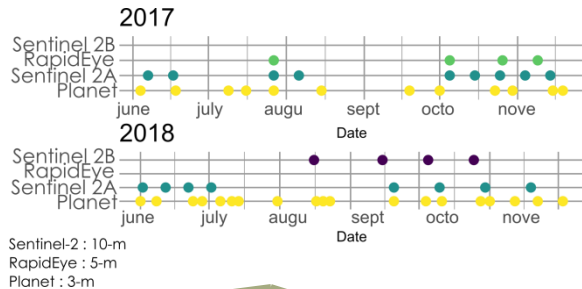


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Linear Mixed model

→ 1-TREE EFFECT

Linear Regression model

→ 2-MILLET YIELD ESTIMATES

WITH TREE

WITHOUT TREE

*Plot level



Where, What, How? Materials & Methods



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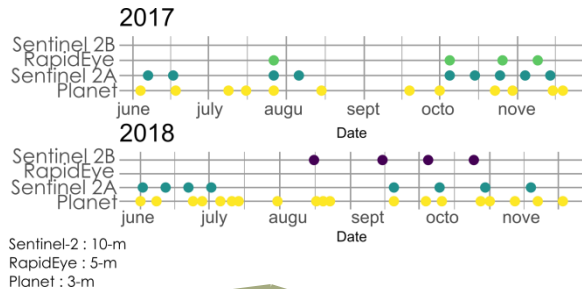


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Gradient Boosting Regression tree model

→ 3-ANALYSIS OF YIELDS HETEROGENEITY

3. Soil information

- *Texture
- *Soil Org Carbon/Soil Org Nitrogen

*Plot level



Results ~ from a ground perspective



Trees effect at landscape scale based on ground observations

Type II Anova with Kenward-Roger ddf approximation for small sample

Variable	Type II F	P-value
Tree density	47.50	<0.001
F.Albida density	3.01	0.09
F.Albida proportion	12.36	<0.001
Tree species richness	31.78	<0.001
kgN.ha ⁻¹	12.69	<0.001
Landscape classe	28.47	<0.001



*Significant effects of parkland on millet yields

Results ~ from a ground perspective



Trees effect at landscape scale based on ground observations

Type II Anova with Kenward-Roger ddf approximation for small sample

Variable	Type II F	P-value	Landscape class	Tree species richness	F.albida proportion	Millet grain yield
Tree density	47.50	<0.001	Class.1	2.5	18	629
F.Albida density	3.01	0.09	Class.4	4	4	911
F.Albida proportion	12.36	<0.001	Class.2	5.6	21	827
Tree species richness	31.78	<0.001	Class.3	1.7	60	1334
kgN.ha ⁻¹	12.69	<0.001	pvalue	0.02	0.11	0.09
Landscape classe	28.47	<0.001				



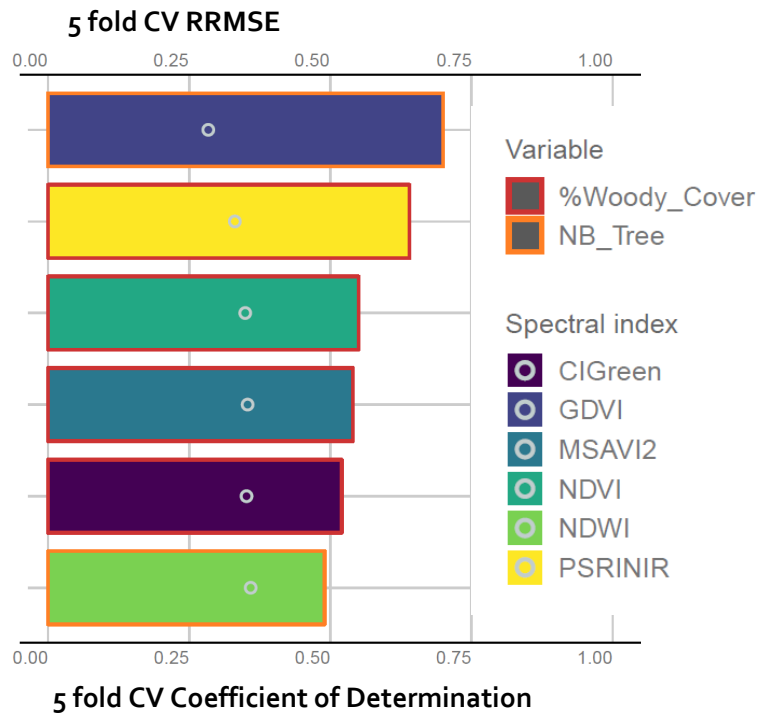
- *Significant effects of parkland on millet yields
- *The well-known 'fertility hotspot' of *F.albida* can be mitigated at landscape scale by the tree species richness and proportion of *F.albida* within fields

Results ~ from a plot perspective



From satellite information to yield estimates accounting for tree effects

1~Sensitivity to vegetation productivity proxy and tree information



*Integrating parklands structuring information improves millet yield model

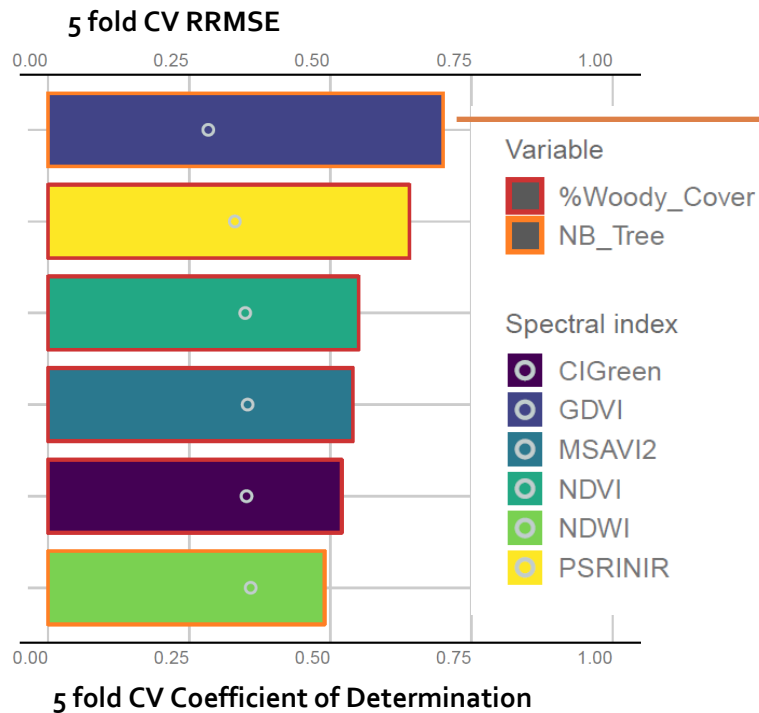
*Best model : GDVI x Nb of trees (R^2 0.70 & RRMSE = 0.28)

Results ~ from a plot perspective

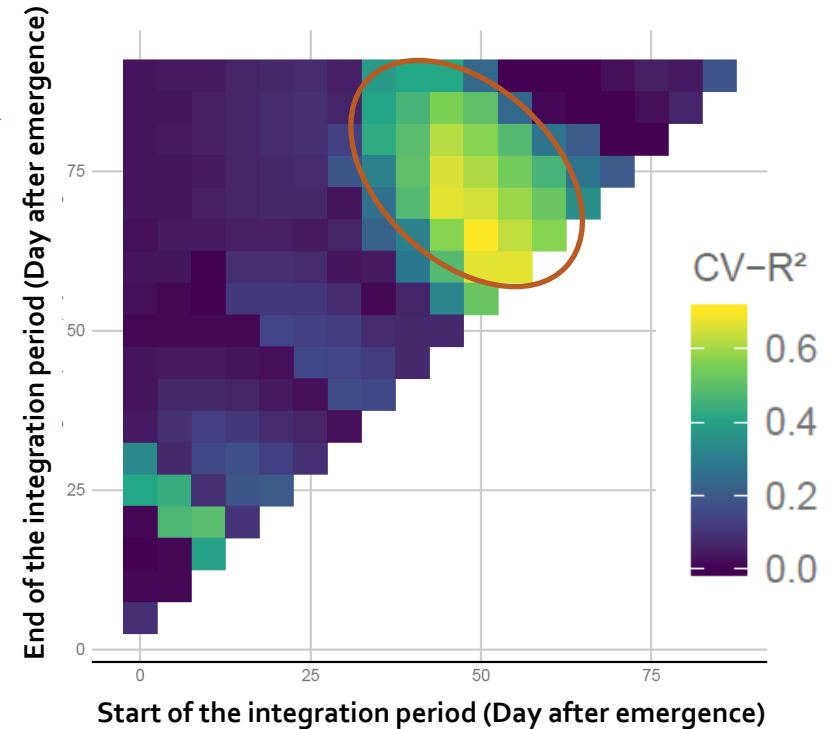


From satellite information to yield estimates accounting for tree effects

1~Sensitivity to vegetation productivity proxy and tree information



2~Sensitivity to phenological development for GDVI



*Integrating parklands structuring information improves millet yield model

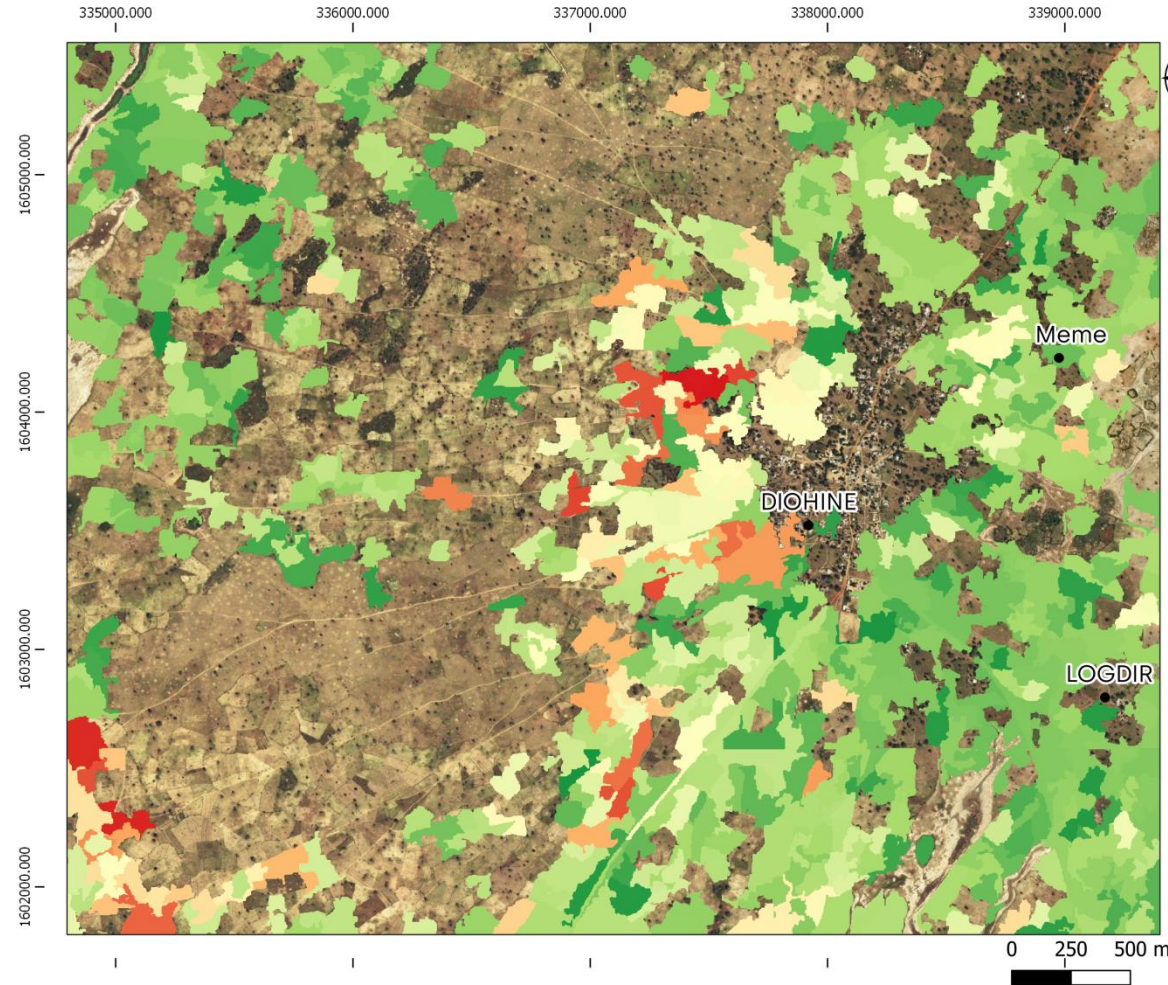
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* Panicle initiation phase to mid of the grain filling phase are more sensitive period

Results ~ from a landscape perspective

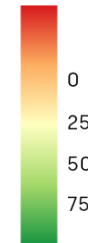


Millet yield heterogeneity analysis at landscape scale

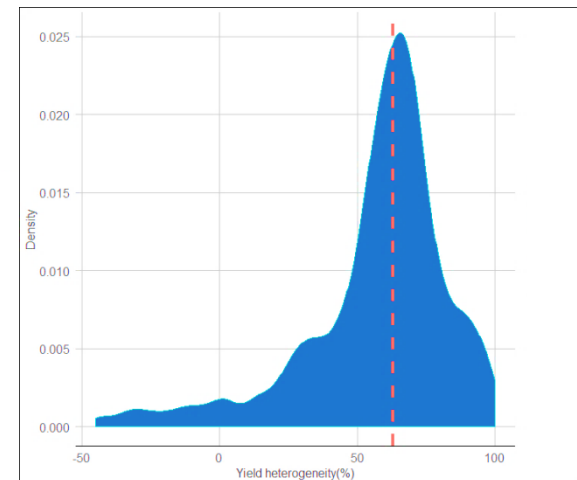


HETEROGENITY IN 2018

Millet heterogeneity
(% of 95th)



- * Comparison with the 95th percentile
- * 95th.p > 1912 kg/ha



- *Median millet yield estimates = 730 kg/ha with high variability (coef.var = 61%)
- *High spatial heterogeneity, with a clear spatial pattern

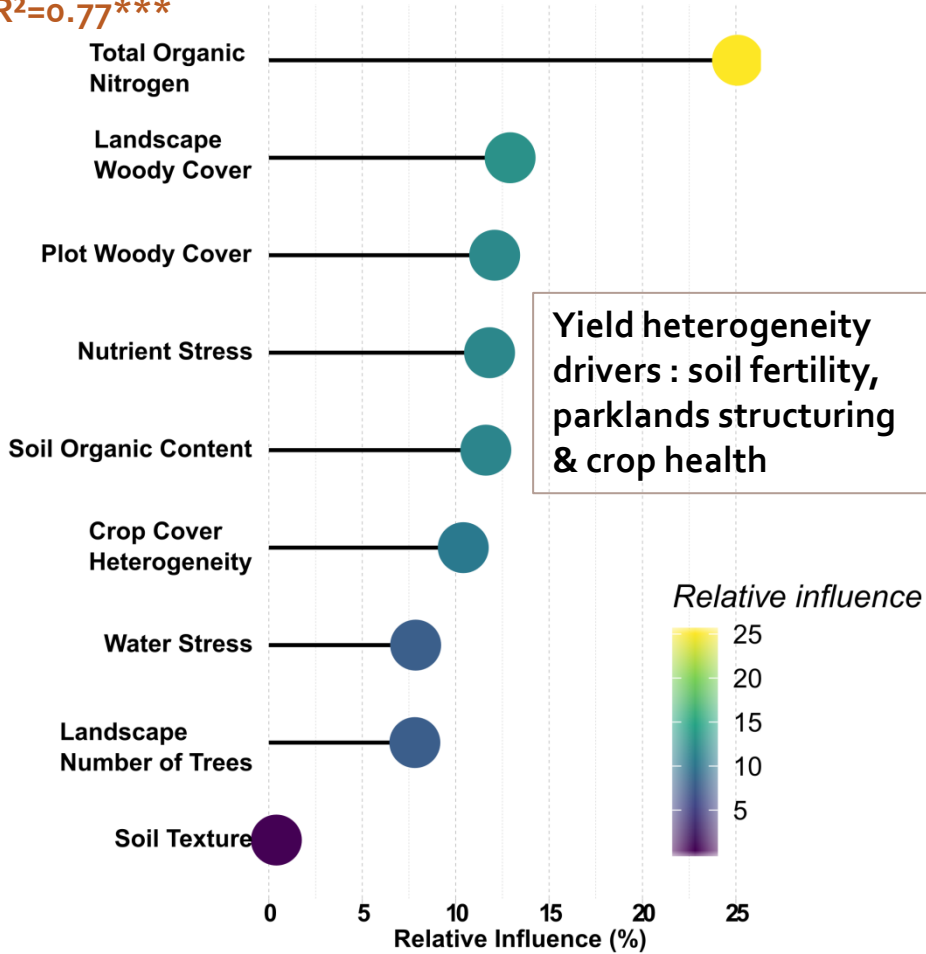
Results ~ from a landscape perspective



What are drivers of spatial heterogeneity pattern?

VARIABLE IMPORTANCE FOR THE GRADIENT BOOSTING TREE

$R^2=0.77***$



*Parkland structuring information and soil fertility as drivers of spatial heterogeneity

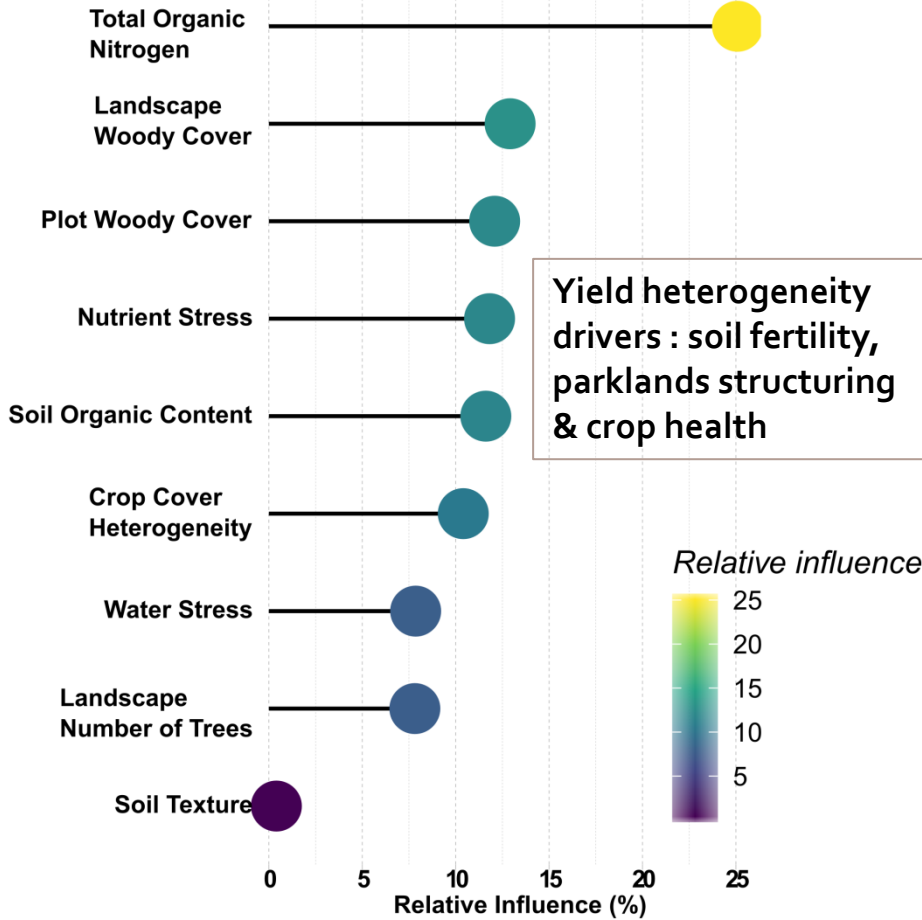
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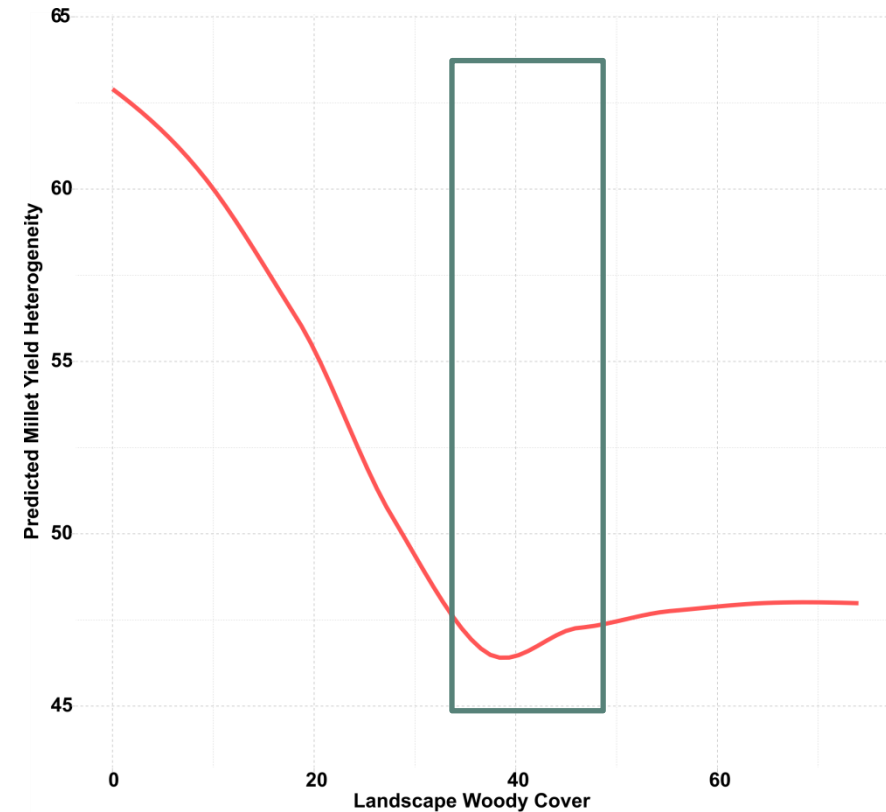
VARIABLE IMPORTANCE FOR THE GRADIENT BOOSTING TREE

$R^2=0.77^{***}$



PARTIAL VARIABLE DEPENDANCE PLOT

Influence of woody cover in surroundig landscape



- *Parkland structuring information and soil fertility as drivers of spatial heterogeneity
- * Woody cover in field surrounding landscape decreases the YH till a certain level

Take home messages





FINDINGS:

- Using parklands structuring information **improves the agricultural performance assessment**
- The **apparent benefits** of individual trees on crop yield can be challenged at landscape scale with the **parklands composition/structuring**



Take home messages



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WHAT IS NEW?

- To **scaling up** crop yields estimates in parklands **using cutting edge multisources remote sensing images**
- To **consider the landscape scale** to explore & improve our understanding on the implication of trees on crop productivity



Take home messages



FINDINGS:

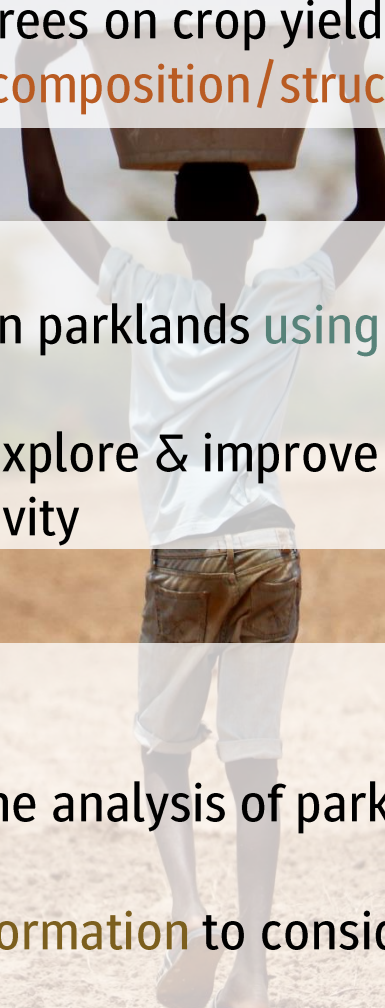
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WHAT IS NEW?

- To **scaling up** crop yields estimates in parklands using cutting edge multisources remote sensing images
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NEXT STEPS?

- To analyze the **intra-field variability**
- To map **tree species** to strengthen the analysis of parklands impacts at landscape scale
- To combine with **socio-economic information** to consider tradeoffs and synergies between goods & services





THANKS FOR YOUR ATTENTION

QUESTIONS-REMARKS : louise.leroux@cirad.fr - <https://louise.leroux.igeo.fr/>

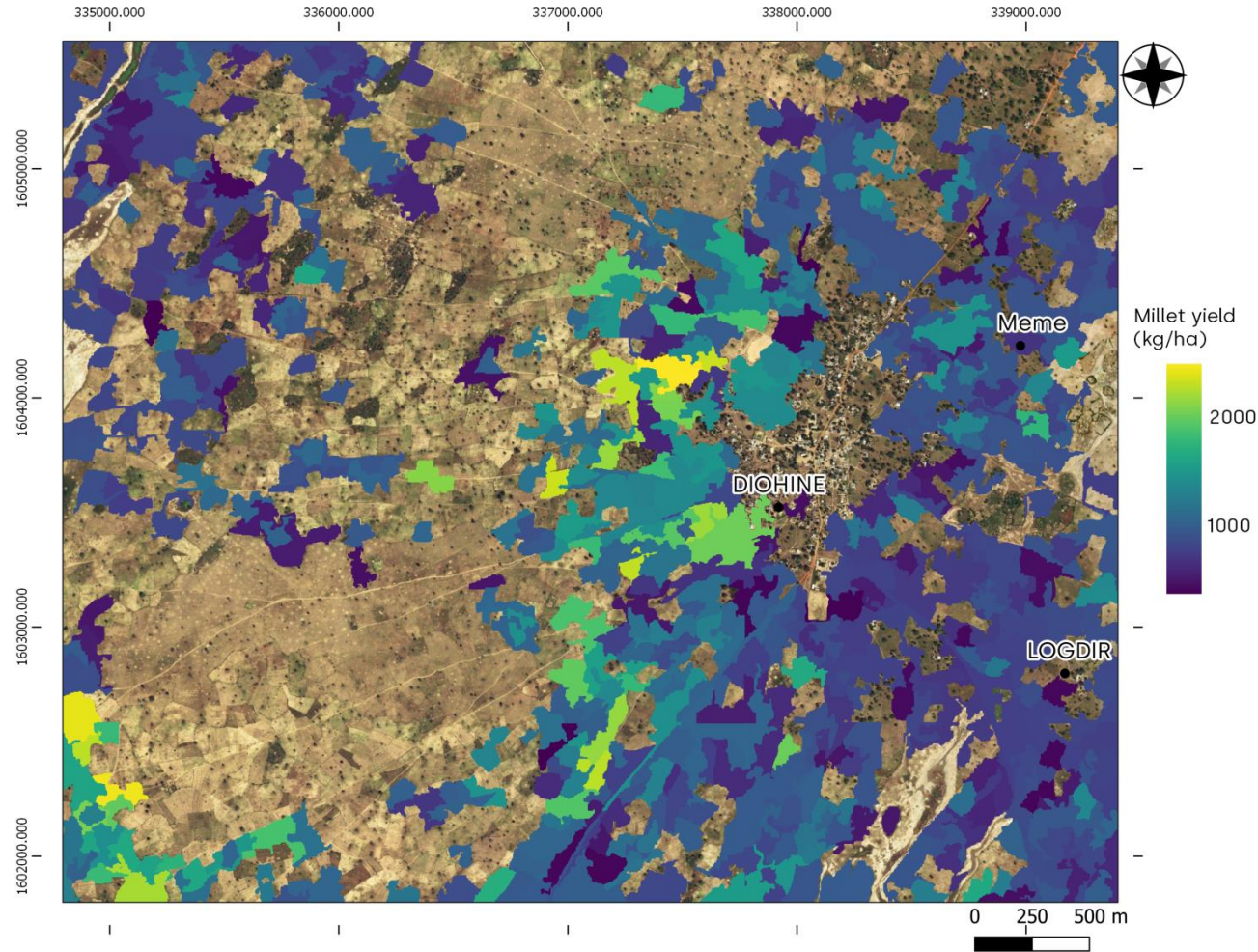
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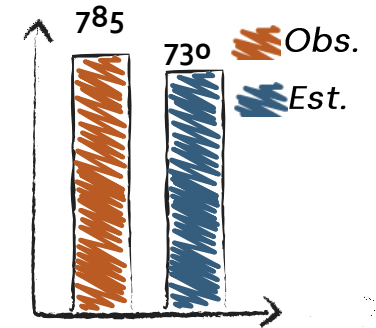
Results ~ from a landscape perspective



Millet yield heterogeneity analysis at landscape scale



MEDIAN YIELD IN 2018



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